- ¹ Title: Tree height and leaf drought tolerance traits shape growth responses across droughts in a temperate
- 2 broadleaf forest

12

- ³ Authors: Ian R. McGregor^{1,2}, Ryan Helcoski¹, Norbert Kunert^{1,3}, Alan J. Tepley^{1,4}, Erika B.
- 4 Gonzalez-Akre¹, Valentine Herrmann¹, Joseph Zailaa^{1,5}, Atticus E.L. Stovall^{1,6,7}, Norman A. Bourg¹,
- ⁵ William J. McShea¹, Neil Pederson⁸, Lawren Sack^{9,10}, Kristina J. Anderson-Teixeira^{1,3*}

6 Author Affiliations:

- Conservation Ecology Center; Smithsonian Conservation Biology Institute; National Zoological Park,
 Front Royal, VA 22630, USA
- 2. Center for Geospatial Analytics; North Carolina State University; Raleigh, NC 27607, USA
- 3. Center for Tropical Forest Science-Forest Global Earth Observatory; Smithsonian Tropical Research Institute; Panama, Republic of Panama
 - 4. Canadian Forest Service, Northern Forestry Centre, Edmonton, Alberta, Canada
- 5. Biological Sciences Department; California State University; Los Angeles, CA 90032, USA
- 6. Department of Environmental Sciences, University of Virginia, Charlottesville, VA 22903, USA
- 7. NASA Goddard Space Flight Center; Greenbelt, MD 20771, USA
- 8. Harvard Forest, Petersham, MA 01366, USA
- 9. Department of Ecology and Evolutionary Biology; University of California, Los Angeles; Los Angeles,
 CA 90095, USA
- 10. Institute of the Environment and Sustainability; University of California, Los Angeles; Los Angeles, CA 90095, USA
- *corresponding author: teixeirak@si.edu; +1 540 635 6546

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Results	697		
Discussion	1467		
Acknowledgements	125		

22 Summary

- As climate change is driving increased drought frequency and severity in many forested regions around
 the world, mechanistic understanding of the factors conferring drought resistance in trees is
 increasingly important. The dendrochronological record provides a window through which we can
 understand how tree size and species' traits shape tree growth responses during droughts.
- We analyzed tree-ring records for twelve species that comprise 97% of the woody productivity of the 25.6-ha ForestGEO plot in a broadleaf deciduous forest of northern Virginia (USA) to test hypotheses on how tree height, microenvironment characteristics, and species' traits shaped drought responses across the three strongest regional droughts over a 60-year period (1950 2009).
- Individual-level drought resistance decreased with tree height, which was strongly correlated with
 exposure to higher evaporative demand and solar radiation. The potentially greater rooting volume of
 larger trees did not confer an advantage in sites with low topographic wetness index. Resistance was
 greater among species whose leaves experienced less shrinkage upon desiccation and lost turgor (wilted)
 at more negative water potentials.
- We conclude that tree height and leaf drought tolerance traits influence growth responses during
 drought, as recorded in the tree-ring record spanning historical droughts. Thus, these factors can be
 useful for predicting future drought responses under climate change.
- Key words: annual growth; crown exposure; drought; Forest Global Earth Observatory (ForestGEO); leaf drought tolerance traits; temperate broadleaf deciduous forest; tree height; tree-ring

41 Introduction

- Forests play a critical global role in climate regulation (?), yet there remains enormous uncertainty as to how the forest-dominated terrestrial carbon sink will respond to climate change (?). An important aspect of this 43 uncertainty lies with physiological responses of trees to drought (?). In many forested regions around the 44 world, the risk of severe drought is increasing (??), often despite increasing precipitation (??). Droughts, 45 intensified by climate change, have been affecting forests worldwide and are expected to continue as one of the most important drivers of forest change in the future (??). Understanding forest responses to drought 47 requires elucidation of how tree size, microenvironment, and species' traits jointly influence individual-level drought resistance, and the extent to which their influence is consistent across droughts. Because the resistance and resilience of growth to drought is linked to trees' probability of surviving drought (??), understanding growth responses can also help elucidate which trees are most vulnerable to drought-induced 51 mortality. However, it has proven difficult to resolve the many factors affecting tree growth during drought with available forest census data, which only rarely captures extreme drought, and with tree-ring records, which capture multiple droughts but usually only sample a subset of a forest community, typically focusing 54 on a single species or the largest individuals. Many studies have shown that within and across species, large trees tend to be more affected by drought. Greater growth reductions for larger trees were first shown on a global scale by ?, and subsequent studies have reinforced this finding (e.g., ??). It has yet to be resolved which of several potential underlying mechanisms most strongly shape these trends in drought response. First, tree height itself may be a primary driver. Taller trees face the biophysical challenge of lifting water greater distances against the effects of gravity and friction (?????). Vertical gradients in stem and leaf traits-including smaller and thicker leaves (higher leaf mass per area, LMA), greater resistance to hydraulic dysfunction (i.e., more negative water potential at 50% loss of hydraulic conductivity, more negative P50), and lower hydraulic conductivity at 63 greater heights (???)—enable trees to become tall (?). Greater stem capacitance (i.e., water storage capacity) of larger trees may also confer resistance to transient droughts [?*; ?]. Indeed, tall trees require xylem of greater hydraulic efficiency, such that xylem conduit diameters are wider in the basal portions of taller trees, both within and across species (??), and throughout the conductive systems of angiosperms (???). Wider 67 xylem conduits plausibly make large trees more vulnerable to embolism during drought (?), and traits conducive to efficient water transport may also lead to poor ability to recover from or re-route water around embolisms (?). 70 Larger trees may also have lower drought resistance because of microenvironmental and ecological factors. Their crowns tend to occupy more exposed canopy positions, which are associated with higher evaporative demand (?). Subcanopy trees tend to fare better specifically due to the benefits of a buffered environment 73 (?). Counteracting the liabilities associated with tall height, large trees tend to have larger root systems (?), potentially mitigating some of the biophysical challenges they face by allowing greater access to water. Larger root systems—if they grant access to deeper water sources—would be particularly advantageous in drier 76 microenvironments (e.g., hilltops, as compared to valleys and streambeds) during drought. Finally, tree size-related responses to drought can be modified by species' traits and their distribution across size classes (??). Understanding the mechanisms driving the greater relative growth reductions of larger trees during 79 drought requires sorting out the interactive effects of height and associated exposure, root water access, and species' traits.
- Debates have also arisen regarding the traits influencing tree growth responses to drought. Studies within

temperate broadleaf forests have observed ring-porous species showing higher drought tolerance than diffuse-porous species (???), but this distinction would not hold in the global context (??) and does not resolve differences among the many species within each category. Commonly-measured traits including wood density and leaf mass per area (LMA) have been linked to drought responses within some temperate deciduous forests (????) and across forests worldwide (?). However, in other cases these traits could not explain drought tolerance (e.g., in a tropical rainforest; ?), or the direction of response was not always consistent. For instance, higher wood density has been associated with greater drought resistance at a global scale (?), but correlated negatively with tree performance during drought in a broadleaf deciduous forest in the southeastern United States (?). Thus, the perceived influence of these traits on drought resistance may actually reflect indirect correlations with other traits that more directly drive drought responses (?). 92 In contrast, hydraulic traits have direct physiological linkages to tree growth and mortality responses to 93 drought. For instance, water potentials at which percent the loss of conductivity surpasses a certain threshold (e.g., P50 and P88, representing 50 and 88% loss of conductivity, respectively) and hydraulic safety margin (i.e., difference between typical minimum water potentials and P50 or P88) correlate with drought performance across global forests (?). However, these are time-consuming to measure and therefore infeasible 97 for predicting or modeling drought responses in highly diverse forests (e.g., in the tropics). More easily-measurable leaf drought tolerance traits that have direct linkage to plant hydraulic function can explain variation in plant distribution and function (?). These include leaf area shrinkage upon desiccation 100 $(PLA_{dry}; ?)$ and the leaf water potential at turgor loss point (π_{tlp}) , i.e., the water potential at which leaf 101 wilting occurs (??). Both traits correlate with hydraulic vulnerability and drought tolerance as part of 102 unified plant hydraulic systems (????). The abilities of both PLA_{dry} and π_{tlp} to explain tree drought 103 resistance remains untested. 104 Here, we examine how tree height, microenvironment characteristics, and species' traits collectively shape 105 drought resistance, defined as the ratio of annual growth in a drought year to that which would be expected 106 in the absence of drought based on previous years' growth. We test a series of hypotheses and associated 107 specific predictions (Table 1) based on the combination of tree-ring records from the three strongest droughts 108 over a 60-year period (1950 - 2009), species trait measurements, and census and microenvironmental data 109 from a large forest dynamics plot in Virginia, USA. First, we focus on how tree size, alone and in its 110 interaction with microenvironmental gradients, influences drought resistance. We examine the contemporary 111 relationship between tree height and microenvironment, including growing season meteorological conditions and crown exposure. We then test whether, consistent with most forests globally, larger-diameter, taller trees 113 tend to have lower drought resistance in this forest, which is in a region (eastern North America) represented 114 by only two studies in the global review of Bennett et al. (2015). We also test for an influence of potential 115 access to available soil water, which should be greater for larger trees in dry but not in perpetually wet 116 microsites. Finally, we focus on the role of species' traits, testing the hypothesis that species' 117 traits--particularly leaf leaf drought tolerance traits--predict drought resistance. We test predictions that 118 drought resistance is higher in ring-porous than semi-ring and diffuse-porous species and that it is correlated 119 with wood density—either positively (?) or negatively (?) and positively correlated with LMA. We further 120 test predictions that species with low PLA_{dry} have higher drought resistance, and that species whose leaves 121

lose turgor lower water potentials (more negative π_{tlp}) have higher resistance.

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Materials and Methods

- 124 Study site and microclimate
- Research was conducted at the 25.6-ha ForestGEO (Forest Global Earth Observatory) study plot at the
- Smithsonian Conservation Biology Institute (SCBI) in Virginia, USA (38°53'36.6"N, 78°08'43.4"W; Fig. S1)
- 127 (??). SCBI is located in the central Appalachian Mountains near the northern boundary of Shenandoah
- National Park. Elevations range from 273 to 338 m above sea level with a topographic relief of 65m (?).
- Climate is humid temperate, with mean annual temperature of 12.7°C and precipitation of 1005 mm yr⁻¹
- during our study period (1960-2009; source: CRU TS v.4.01; ?). Dominant tree taxa within this secondary
- 131 forest include Liriodendron tulipifera, oaks (Quercus spp.), and hickories (Carya spp.; Table 2).
- 132 Identifying drought years
- 133 We identified the three largest droughts within the time period 1950-2009, defining drought (?) as events
- with anomalously dry peak growing season climatic conditions. Specifically, we used the metric of Palmer
- Drought Severity Index (PDSI) during May-August (MJJA; Table S1), which were identified by ? as the
- months of the current year to which annual tree growth was most sensitive at this site. PDSI divisional data
- 137 for Northern Virginia were obtained from NOAA
- 138 (https://www7.ncdc.noaa.gov/CDO/CDODivisionalSelect.jsp) in December 2017. Based on this, we
- 139 identified the three strongest droughts during the study period (Figs. 1, S1; Table S1).
- The droughts differed in intensity and antecedent moisture conditions (Fig. S1, Table S1). The 1966 drought
- was preceded by two years of moderate drought during the growing season and severe to extreme drought
- starting the previous fall. In August 1966, PDSI reached its lowest monthly value (-4.82) of the three
- droughts. The 1977 drought was the least intense throughout the growing season, and it was preceded by 2.5
- years of near-normal conditions, making it the mildest of the three droughts. The 1999 drought was preceded
- by wetter than average conditions until the previous June, but PDSI plummeted below -3.0 in October 1998
- and remained below this threshold through August 1999.
- 147 Data collection and preparation
- Within or just outside the ForestGEO plot, we collected data on a suite of variables including tree heights,
- microenvironment characteristics, and species traits (Table 3). The SCBI ForestGEO plot was censused in
- $_{150}$ 2008, 2013, and 2018 following standard ForestGEO protocols, whereby all free-standing woody stems ≥ 1 cm
- diameter at breast height (DBH) were mapped, tagged, measured at DBH, and identified to species (?).
- 152 From these census data, we used measurements of DBH from 2008 to calculate historical DBH and data for
- 153 all stems ≥ 10cm to analyze functional trait composition relative to tree height (all analyses described
- below). Census data are available through the ForestGEO data portal (www.forestgeo.si.edu).
- We analyzed tree-ring data (xylem growth increment) from 571 trees representing the twelve dominant
- species (Table 2; Fig. S2). Selected species were those with the greatest contributions to woody aboveground
- net primary productivity $(ANPP_{stem})$ and together comprised 97% of study plot $ANPP_{stem}$ between 2008
- and 2013 (?). Cores (one per tree) were collected within the ForestGEO plot at breast height (1.3m) in
- 2010-2011 or 2016-2017. In 2010-2011, cores were collected from randomly selected live trees of each species
- that had at least 30 individuals \geq 10 cm DBH (?). Annual tree mortality censuses were initiated in 2014 (?),
- and in 2016-2017, cores were collected from all trees found to have died since the previous year's census. We
- $_{162}$ note that drought was probably not a cause of mortality for these trees, as monthly May-Aug PDSI did not

drop below -1.75 in these years or the three years prior (2013-2017), and that trees cored dead displayed similar climate sensitivity to trees cored live (?). Cores were sanded, measured, and crossdated using standard procedures, as detailed in (?). The resulting chronologies (Fig. 1a) were published in Zenodo (DOI: 10.5281/zenodo.2649302) in association with ?.

For each cored tree, we combined tree-ring records and allometric equations of bark thickness to reconstruct DBH for the years 1950-2009. Prior *DBH* was estimated using the following equation:

$$DBH_Y = DBH_{2008} - 2 * \left[r_{bark,2008} - r_{bark,Y} + \sum_{year=Y}^{2008} r_{ring,Y} \right]$$

Here, Y denotes the year of interest, r_{ring} denotes ring width derived from cores, and r_{bark} denotes bark

thickness. Bark thickness was estimated from species-specific allometries based on the bark thickness data 170 from the site (?). Specifically, we used linear regression on log-transformed data to relate r_{bark} to diameter 171 inside bark from 2008 data (Table S2), which were then used to determine r_{bark} in the DBH reconstruction. 172 Tree heights (H) were measured by several researchers for a variety of purposes between 2012 and 2019 173 (n=1,518 trees). Methods included direct measurements using a collapsible measurement rod on small trees 174 (?) or a tape measure on recently fallen trees (this study); geometric calculations using clinometer and tape 175 measure (?) or digital rangefinders (??); and ground-based LiDAR (?). Rangefinders used either the tangent method (Impulse 200LR, TruPulse 360R) or the sine method (Nikon ForestryPro) for calculating heights. 177 Both methods are associated with some error (?), but in this instance there was no clear advantage of one or 178 the other. Measurements from the National Ecological Observatory Network (NEON) were collected near the 179 ForestGEO plot following standard NEON protocol, whereby vegetation of short stature was measured with 180 a collapsible measurement rod, and taller trees with a rangefinder (?). Species-specific height allometries 181 were developed (Table S3) using log-log regression $(ln[H] \sim ln[DBH])$. For species with insufficient height 182 data to create reliable species-specific allometries (n=2, JUNI and FRAM), heights were calculated from an 183 equation developed by combining the height measurements across all species. We then used these allometries to estimate H for each drought year, Y, based on reconstructed DBH_Y . The distribution of H across 185 drought years is shown in Fig. S3. 186

To characterize how environmental conditions vary with height, data were obtained from the NEON tower located <1km from the study area via the neonUtilities package (?). We used wind speed, relative humidity, and air temperature data, all measured over a vertical profile spanning heights from 7.2 m to above the top of the tree canopy (31.0 or 51.8m, depending on censor), for the years 2016-2018 (?). After filtering for missing and outlier values, we determined the daily minima and maxima, which we then aggregated at the monthly scale.

Crown position—a categorical variable classifying trees based on exposure to sunlight—was recorded for all cored trees that remained standing during the growing season of 2018 following the protocol of ?. Trees were classified as follows: dominant trees were defined as those with crowns above the general level of the canopy, co-dominant trees as those with crowns within the tanopy; intermediate trees as those with crowns below the canopy level, but illuminated from above; and suppressed as those below the canopy and receiving minimal direct illumination from above.

⁹⁹ Topographic wetness index (TWI), used here as a metric of long-term mean moisture availability, was

a hydrological run-off model and has since been used for a number of purposes in hydrology and ecology (?). TWI calculation depends on an input of a digital elevation model (DEM; ~3.7 m resolution from the elevatr 202 package (?)), and from this yields a quantitative assessment defined by how "wet" an area is, based on areas 203 where run-off is more likely. From our observations in the plot, TWI performed better at categorizing wet areas than the Euclidean distance from the stream. 205 Species' trait data were collected in August 2018 (Tables 2-3; Fig. S4). We sampled small, sun-exposed 206 branches up to eight meters above the ground from three individuals of each species in and around the 207 ForestGEO plot. Sampled branches were re-cut under water at least two nodes above the original cut and re-hydrated overnight in covered buckets under opaque plastic bags before measurements were taken. 209 Rehydrated leaves taken towards the apical end of the branch (n=3 per individual: small, medium, and 210 large) were scanned, weighed, dried at 60° C for ≥ 48 hours, and then re-scanned and weighed. Leaf area was calculated from scanned images using the LeafArea R package (?). LMA was calculated as the ratio of 212 leaf dry mass to fresh area. PLA_{dry} was calculated as the percent loss of area between fresh and dry leaves. 213 Wood density was calculated for ~1cm diameter stem samples (bark and pith removed) as the ratio of dry 214 weight to fresh volume, which was estimated using Archimedes' displacement. We used the rapid 215 determination method of ? to estimate osmotic potential at turgor loss point (π_{tlp}) . Briefly, two 4 mm 216 diameter leaf discs were cut from each leaf, tightly wrapped in foil, submerged in liquid nitrogen, perforated 217 10-15 times with a dissection needle, and then measured using a vapour pressure osmometer (VAPRO 5520, 218 Wescor, Logan, UT, USA). Osmotic potential (π_{osm}) given by the osmometer was used to estimate (π_{tlp}) 219 using the equation $\pi_{tlp} = 0.832\pi_{osm}^{-0.631}$ (?). Statistical Analysis 221 For each drought year, we calculated a metric drought resistance (Rt) as the ratio of basal area increment 222 (BAI; i.e., change in cross-sectional area) during the drought year to the mean BAI over the five years 223 preceding the drought (?). Thus, Rt values <1 and >1 indicate growth reductions and increases, respectively. 224 Because the Rt metric could be biased by directional pre-drought growth trends, we also tried an 225 intervention time series analysis (ARIMA, (?)) that predicted mean drought-year growth based on trends 226 over the preceding ten years and used this value in place of the five-year mean in calculations of resistance 227 $(Rt_{ARIMA} = \text{observed } BAI/\text{ predicted } BAI)$. The two metrics were strongly correlated (Fig. S5). Visual 228 review of the individual tree-ring sequences with the largest discrepancies between these metrics revealed 229 that Rt was less prone to unreasonable estimates than Rt_{ARIMA} , so we selected Rt as our focal metric, presenting parallel results for Rt_{ARIMA} in the Supplementary Info. In this study we focus exclusively on 231 drought resistance metrics (Rt or Rt_{ARIMA}), and not on the resilience metrics described in ?, because (1) we 232 would expect resilience to be controlled by a different set of mechanisms, and (2) the findings of (?) suggest that Rt is a more important drought response metric for angiosperms in that low resistance to moderate 234 droughts was a better predictor of mortality during subsequent severe droughts than the resilience metrics. 235 Analyses focused on testing the predictions presented in Table 1 with Rt as the response variable, and then 236 repeated using Rt_{ARIMA} as the response variable. Models were run for all drought years combined and for 237 each drought year individually. The general statistical model for hypothesis testing was a mixed effects 238

calculated using the dynatopmodel package in R (Fig. S2) (?). Originally developed by ?, TWI was part of

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model, implemented in the lme4 package in R (?). In the multi-year model, we included a random effect of

tree nested within species and a fixed effect of drought year to represent the combined effects of differences in

drought characteristics. Individual year models included a random effect of species. All models included fixed

AICc refers to a corrected version of AICc, and is best suited for small data sizes (see ?). 245 To avoid over-fitting models with five species traits (Table 3) across only 12 species, we did not include all 246 traits as fixed effects in a single linear mixed model, but rather conducted individual tests of each species trait to determine the relative importance and appropriateness for inclusion in the main model. These tests 248 followed the model structure specified above, then added ln[H] and ln[TWI] to create a base model against 249 which we tested traits. Trait variables were considered appropriate for inclusion in the main model if they had a consistent direction of response across all droughts and if their addition to the base model improved fit 251 (at $\triangle AICc \ge 1.0$) in at least one drought year (Table S4). We note that we did not use the $\triangle AICc \ge 1.0$ 252 criterion as a test of significance, but rather of whether the variable had enough influence to be considered as a candidate variable in full models. 254

effects of independent variables of interest (Tables 1,3) as specified below. All variables across all best models

conditional/marginal R-squared to assess model fit as implemented in the AICcmodayg package in R (?).

had variance inflation factors <1.2 (1 +/- 0.019). We used AICc to assess model selection, and

We then determined the top full models for predicting Rt (or Rt_{ARIMA}). To do so, we compared models 255 with all possible combinations of candidate variables, including $ln[H]^*ln[TWI]$ and species traits as specified 256 above. We identified the full set of models within $\triangle AICc=2$ of the best model (that with lowest AICc). 257 When a variable appeared in all of these models and the sign of the coefficient was consistent across models, 258 we viewed this as support for the acceptance/rejection of the associated prediction (Table 1). If the variable 259 appeared in some but not all of these models, and its sign was consistent across models, we considered this 260 partial support/rejection. In presentation of the results below, we note instances where the Rt_{ARIMA} model 261 disagreed with the Rt model, but otherwise do not discuss the Rt_{ARIMA} model. 262

All analysis beyond basic data collection was performed using R version 3.6.2 (?). Other R-packages used in analyses are listed in the Supplementary Information (*Appendix S1*). All data, code, and results are available through the SCBI-ForestGEO organization on GitHub (https://github.com/SCBI-ForestGEO:

SCBI-ForestGEO-Data and McGregor_climate-sensitivity-variation repositories), with static versions corresponding to data and analyses presented here archived in Zenodo (DOIs: 10.5281/zenodo.3604993 and /TBD), respectively.

269 Results

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270 Tree height and microenvironment

In the years for which we have vertical profiles in climate data (2016-2018), taller trees—or those in dominant crown positions— were generally exposed to higher evaporative demand during the peak growing season months (May-August; Fig. 2). Specifically, maximum daily wind speeds were significantly higher above the top of the canopy (40-50m) than within and below (10-30m) (Fig. 2a). Relative humidity was also somewhat lower during June-August, ranging from ~50-80% above the canopy and ~60-90% in the understory (Fig. 2b).

Air temperature did not vary consistently across the vertical profile (Fig. 2c).

277 Crown position varied as expected with height (dominant > co-dominant > intermediate > suppressed), but 278 with substantial variation (Fig. 2d). There were significant differences in height across all crown position 279 classes (Fig. 2d). A comparison test between height and crown position data from the most recent 280 ForestGEO census (2018) revealed a correlation of 0.73.

Community-level drought responses 281 At the community level, cored trees showed substantial growth reductions in all three droughts, with a mean 282 Rt of 0.86 in 1966 and 1999, and 0.84 in 1977 (Fig. 2b). Across the entire study period (1950-2009), the 283 focal drought years were the three years with the largest fraction of trees exhibiting Rt < 0.7. Specifically, in 284 each drought, roughly 30% of the cored trees had growth reductions of > 30% (Rt < 0.7): 29% in 1966, 32% 285 in 1977, and 27% in 1999. However, some individuals exhibited increased growth, i.e., Rt > 1.0: 26% of trees in 1966, 22% in 1977, and 26% in 1999. 287 In the context of the multivariate model, Rt did not vary across drought years. That is, drought year as a 288 variable did not appear in any of the top models -i.e., models that were statistically indistinguishable 289 $(\Delta AICc < 2)$ from the best model. Tree height, microenvironment, and drought resistance 291 Taller trees (based on H in the drought year) showed stronger growth reductions during drought (Table 1; 292 Figs. 4, S6). Specifically, ln[H] appeared, with a negative coefficient, in the best model (($\Delta AICc=0$) and all 293 top models when evaluating the three drought years together (Tables S6-S7). The same held true for 1966 individually. For the 1977 drought, ln[H] did not appear in the best model, but was included, with a 295 negative coefficient, among the top models-i.e., models that were statistically indistinguishable ($\triangle AICc<2$) 296 from the best model (Tables 1, S6-S7). For the 1999 drought, ln[H] had no significant effect. Rt had a significantly negative response to ln[TWI] across all drought years combined (Figs. 4, S6, Table S6-S7). The effect was also significant for 1977 and 1999 individually (Fig. 4, Table S6). When Rt_{ARIMA} 299 was used as the response variable, the effect was significant in 1977, and included in some of the top models 300 in 1966 and 1999 (Table S7). This negates the idea that trees in moist microsites would be less affected by drought. Nevertheless, we tested for a ln[H] * ln[TWI] interaction, a negative sign of which could indicate 302 that smaller trees (presumably with smaller rooting volume) are more susceptible to drought in drier 303 microenvironments with a deeper water table. This hypothesis was rejected, as the ln[H] * ln[TWI]interaction was never significant, and had a positive sign in any top models in which it appeared (Tables 1, 305 S6-S7). This term did appear with a positive coefficient in the best Rt_{ARIMA} model for all years combined 306 (Table S7), indicating that the negative effect of height on Rt was significantly stronger in wetter 307 microhabitats. 308 Species' traits and drought resistance 309 Species, as a factor in ANOVA, had significant influence (p<0.05) on all traits (wood density, LMA, 310 PLA_{dry} , and π_{tlp}), with more significant pairwise differences for wood density and PLA_{dry} than for LMA311 and π_{tlp} (Table 2, Fig. S4). Drought resistance also varied across species, overall and in each drought year 312 (Fig. 3). Significant differences in Rt across species were most pronounced in 1966 with a total of seven 313 distinct groupings, while 1977 had four and 1999 had two. Averaged across all droughts, Rt was lowest in Liriodendron tulipifera (mean Rt = 0.66) and highest in Fagus grandifolia (mean Rt = 0.99). 315 Wood density, LMA, and xylem porosity were all poor predictors of Rt (Tables 1,S4-S5). Wood density and 316 LMA were never significantly associated with Rt in the single-variable tests and were therefore excluded 317 from the full models. Xylem porosity was also excluded from the full models, as it had no significant

ring-porous species had higher Rt than diffuse- and semi-ring- porous species in the 1966 and 1999 droughts,

influence for all droughts combined and had contrasting effects in the individual droughts: whereas

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Liriodendron tulipifera and Fagus grandifolia, were at opposite ends of the Rt spectrum (Fig. 3), further refuting the idea that xylem porosity is a useful predictor of Rt in the context of this study. 323 In contrast, PLA_{dry} , and π_{tlp} were both negatively correlated to drought resistance (Figs. 4, S6; Tables 324 1,S4-S7). Both had consistent signs across all droughts, and their inclusion at least marginally improved the 325 model ($\Delta AICc > 1.0$) for at least one of the three droughts (Table S4), qualifying them as candidate variables for the full model. PLA_{dry} had a significant influence, with negative coefficient, in full models for 327 the three droughts combined and for the 1966 drought individually (Fig. 4; Tables S6-S7). For 1977 and 328 1999, it was included with a negative coefficient in some of the top models (Tables S6-S7). π_{tlp} was included with a negative coefficient in the best model for both all droughts combined and for the 1977 drought 330 individually (Fig. 4; Table S6). It was also included in some of the top models for 1999 (Tables S6-S7). 331

they had lower Rt in 1977 (Table S4). It is noteworthy that the two diffuse-porous species in our study,

332 Discussion

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Tree height, microenvironment, and leaf drought tolerance traits shaped tree growth responses across three 333 droughts at our study site (Table 1, Fig. 4). The greater susceptibility of larger trees to drought, similar to forests worldwide (?), was driven primarily by their height (?). Taller height was likely a liability in itself, 335 and was also associated with greater exposure to conditions that would promote water loss and heat damage 336 during drought (Fig. 2). There was no evidence that greater availability of, or access to, soil water 337 availability increased drought resistance; in contrast, trees in wetter topographic positions had lower Rt (??), 338 and the larger potential rooting volume of large trees provided no advantage in the drier microenvironments. 339 The negative effect of height on Rt held after accounting for species' traits, which is consistent with recent work finding height had a stronger influence on mortality risk than forest type during drought (?). Drought 341 resistance was not consistently linked to species' LMA, wood density, or xylem type (ring- vs. diffuse 342 porous), but was negatively correlated with leaf drought tolerance traits (PLA_{dry}, π_{tlp}) . This is the first study to our knowledge linking PLA_{dry} and π_{tlp} to growth reduction during drought. The directions of these 344 responses were consistent across droughts (Table S6), supporting the premise that they were driven by 345 fundamental physiological mechanisms. However, the strengths of each predictor varied across droughts (Fig. 4; Tables S6-S7), indicating that drought characteristics interact with tree size, microenvironment, and traits 347 to shape which individuals are most affected. These findings advance our knowledge of the factors that make 348 trees vulnerable to growth declines during drought and, by extension, likely make them more vulnerable to mortality (?). 350 The droughts considered here were of a magnitude that has occurred with an average frequency of 351 approximately once every 10-15 years (Fig. 1a,?) and had substantial but not devastating impacts on tree growth (Figs. 1b). These droughts were classified as severe (PDSI < -3.0; 1977) or extreme (PDSI < -4.0; 353 1966, 1999) at our site and have been linked to tree mortality in the eastern United States (?). However, 354 extreme, multiannual droughts such as the so-called "megadroughts" of this type that have triggered massive 355 tree die-off in other regions (e.g., ??) have not occurred in the Eastern United States within the past several 356 decades (?). Of the droughts considered here, the 1966 drought, which was preceded by two years of dry 357 conditions (Fig. S1), severely stressed a larger portion of trees (Fig. 1b). The tendency for large trees to 358 have lowest resistance was most pronounced in this drought, consistent with other findings that this 359 physiological response increases with drought severity (??). Across all three droughts, the majority of trees 360 experienced reduced growth, but a substantial portion had increased growth (Fig. 1b), potentially due to

decreased leaf area of competitors during the drought (REF-if we can find one), and consistent with prior observations that smaller trees can exhibit increased growth rates during drought (?). It is likely because of the moderate impact of these droughts, along with other factors influencing tree growth (e.g., 364 stand dynamics), that our best models characterize only a modest amount of variation in Rt: 11-12\% for all 365 droughts combined, and 18-25% for each individual drought (Fig. S6; Table S6). Consistent with studies in other forests worldwide (?), taller trees in this forest exhibited lower drought resistance. Mechanistically, this is consistent with, and reinforces, previous findings that biophysical 368 constraints make it impossible for trees to efficiently transport water to great heights and simultaneously 369 maintain strong resistance and resilience to drought-induced embolism (???). Taller trees also face dramatically different microenvironments (Fig. 2). They are exposed to higher wind speeds and lower 371 humidity (Fig. 2a-b), resulting in higher evaporative demand. Unlike other temperate forests where 372 modestly cooler understory conditions have been documented (?), particularly under drier conditions (?), we observed no significant variation in air temperatures across the vertical profile (Fig. 2c). More critically for 374 tree physiology, leaf temperatures can become significantly elevated over air temperature under conditions of 375 high solar radiation and low stomatal conductance (??). Under drought, when air temperatures tend to be 376 warmer, direct solar radiation tends to be higher (because of less cloud cover), and less water is available for 377 evaporative cooling of the leaves, trees with sun-exposed crowns may not be able to simultaneously maintain 378 leaf temperatures below damaging extremes and avoid drought-induced embolism. Indeed, previous studies 379 have shown lower drought resistance in more exposed trees (???). Unfortunately, collinearity between height 380 and crown exposure in this study (Fig. 2d) makes it impossible to confidently partition causality. Additional 381 research comparing drought responses of early successional and mature forest stands, along with short and 382 tall isolated trees, would be valuable for more clearly disentangling the roles of tree height and crown 383 exposure. 384 Belowground, taller trees would tend to have larger root systems (?), but this does not necessarily imply that 385 they have greater access to or reliance on deep soil-water resources that may be critical during drought. 386 While tree size can correlate with the depth of water extraction (?), the linkage is not consistent. Shorter 387 trees can vary broadly in the depth of water uptake (?), and larger trees may allocate more to abundant 388 shallow roots that are beneficial for taking up water from rainstorms (?). Moreover, reliance on deep 389 soil-water resources can actually prove a liability during severe and prolonged drought, as these can 390 experience more intense water scarcity relative to non-drought conditions (?). In any case, the potentially 391 greater access to water did not override the disadvantage conferred by height-and, in fact, greater moisture 392 access in non-drought years (here, higher TWI) appears to make trees more sensitive to drought (??). This 393 may be because moister habitats would tend to support species and individuals with more mesophytic traits (???), potentially growing to greater heights (e.g., ?), and these are then more vulnerable when drought hits. 395 The observed height-sensitivity of Rt, together with the lack of conferred advantage to large stature in drier 396 topographic positions, agrees with the concept that physiological limitations to transpiration under drought shift from soil water availability to the plant-atmosphere interface as forests age (?), such that tall, dominant 398 trees are the most sensitive in mature forests. Again, additional research comparing drought responses across 399 forests with different tree heights and water availability would be valuable for disentangling the relative 400 importance of above- and belowground mechanisms across trees of different size. 401 The development of tree-ring chronologies for the twelve most dominant tree species at our site (??) gave us 402 the sample size to compare historical drought responses across species (Fig. 3) and associated traits at a

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single site (see also?). Our study reinforced current understanding (see Introduction) that wood density and
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    LMA are not reliably linked to drought resistance (Table 1). Contrary to previous studies in temperate
    deciduous forests, we did not find an association between xylem porosity and drought tolerance, as the two
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    diffuse-porous species, Liriodendron tulipifera and Faqus grandifolia, were at opposite ends of the Rt
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    spectrum (Fig. 3). While the low Rt of L, tulipifera is consistent with other studies (?), the high Rt of F.
    grandifolia contrasts with studies identifying diffuse porous species in general (??), and the genus Fagus in
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    particular (?), as drought sensitive. There are two potential explanations for this discrepancy. First, other
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    traits can and do override the influence of xylem porosity on drought resistance. Ring-porous species are
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    restricted mainly to temperate deciduous forests (?), while highly drought-tolerant diffuse-porous species
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    exist in other biomes (REFS). Fagus grandifolia had intermediate \pi_{tlp} and low PLA_{dry} (Fig. S4), which
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    would have contributed to its drought resistance (Fig. 4; see discussion below). A second explanation of why
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    F. grandifolia trees at this particular site had higher Rt is that the sampled individuals, reflective of the
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    population within the plot, are generally shorter and in less-dominant canopy positions compared to most
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    other species (Fig. S4). The species, which is highly shade-tolerant, also has deep crowns (?), implying that
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    a lower proportion of leaves would be affected by harsher microclimatic conditions at the top of the canopy
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    under drought (Fig. 2). Thus, the high Rt of the sampled F. grandifolia population can be explained by a
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    combination of fairly drought-resistant leaf traits, shorter stature, and a buffered microenvironment.
    Concerted measurement of tree-rings and leaf drought tolerance traits of emerging importance (???) allowed
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    novel insights into the role of drought tolerance traits in shaping drought response. The finding that PLA_{dry}
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    and \pi_{tlp} can be useful for predicting drought responses of tree growth (Fig. 4; Table 1) is both novel and
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    consistent with previous studies linking these traits to habitat and drought tolerance. Previous studies have
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    demonstrated that \pi_{tlp} and PLA_{dry} are physiologically meaningful traits linked to species distribution along
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    moisture gradients (??????), and our findings indicate that these traits also influence drought responses.
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    Furthermore, the observed linkage of \pi_{tlp} to Rt in this forest aligns with observations in the Amazon that
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    \pi_{tlp} is higher in drought-intolerant than drought-tolerant plant functional type. Further, it adds support to
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    the idea that this trait is useful for categorizing and representing species' drought responses in models (?).
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    Because both PLA_{dry} and \pi_{tlp} can be measured relatively easily (??), they hold promise for predicting
    drought growth responses across diverse forests. The importance of predicting drought responses from
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    species traits increases with tree species diversity; whereas it is feasible to study drought responses for all
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    dominant species in most boreal and temperate forests (e.g., this study), this becomes difficult to impossible
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    for species that do not form annual rings, and for diverse tropical forests. Although progress is being made
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    for the tropics (?), a full linkage of drought tolerance traits to drought responses would be invaluable for
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    forecasting how little-known species and whole forests will respond to future droughts (??).
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    As climate change drives increasing drought in many of the world's forests (??), the fate of forests and their
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    climate feedbacks will be shaped by the biophysical and physiological drivers observed here. Our results,
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    consistent with other observations around the world, imply that the tallest, most exposed trees will be most
    affected (??). We show that, at least within the mature forest studied here, the vulnerability conferred by
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    tall height and associated crown exposure outweigh any advantage of a larger root system, even in drier
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    microenvironments. This would suggest that the drought responses of trees in mature forests are more
    strongly differentiated along the size spectrum by their above- than below-ground environment. The same
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    may not be true of systems where short trees exist outside of a buffered understory environment-i.e., open
    grown trees or short-statured, early-successional forests. The latter appear to be limited more strongly by
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root water access during drought (?), and would also be dominated by species with different traits. The
earlier-successional species at our site (*Liriodendron tulipifera*, *Quercus spp.*, *Fraxinus americana*) display a
mix of traits conferring drought tolerance and resistance (Table 2), while the late-successional *Fagus*grandifolia displayed high drought resistance, in part because it exists primarily within a buffered
microenvironment. Further research on how leaf drought tolerance traits and drought vulnerability change
over the course of succession would be valuable for addressing how drought tolerance changes as forests age
(e.g. ?). In the meantime, the results of this study advance our knowledge of the factors conferring drought
resistance in a mature forest, opening the door for more accurate forecasting of forest responses to future
drought.

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466 Author Contribution

KAT, IM, and AJT designed the research. Tree-ring chronologies were developed by RH under guidance of AJT and NP. Trait data were collected by IM, JZ under guidance of NK and LS. Other plot data were collected by IM, AS, EGA, and NB under guidance of EGA and WM. Data analyses were performed by IM under guidance of KAT and VH. KAT and IM interpreted the results. IM and KAT wrote the first draft of manuscript, and all authors contributed to revisions.

472 Supplementary Information

- Table S1: Monthly Palmer Drought Severity Index (PDSI), and its rank among all years between 1950 and 2009 (driest=1), for focal droughts.
- Table S2: Species-specific bark thickness regression equations.
- ⁴⁷⁶ Table S3: Species-specific height regression equations.
- Table S4. Individual tests of species traits as drivers of drought resistance, where Rt is used as the response variable.
- Table S5. Individual tests of species traits as drivers of drought resistance, where Rt_{ARIMA} is used as the response variable.
- Table S6. Summary of top full models for each drought instance, where Rt is used as the response variable.

- Table S7. Summary of top models for each drought instance, where Rt_{ARIMA} is used as the response
- 483 variable.
- Figure S1. Time series of Palmer Drought Severity Index (PDSI) for the 2.5 years prior to each focal drought
- Figure S2: Map of ForestGEO plot showing topographic wetness index and location of cored trees
- Figure S3: Distribution of reconstructed tree heights across drought years.
- Figure S4. Distribution of independent variables by species.
- Figure S5. Comparison of Rt and Rt_{ARIMA} results, with residuals, for each drought scenario
- Figure S6. Visualization of best model, with data, for all droughts combined.